

## Last Time: Orthogonality.

### Gram-Schmidt Process:

Given lin. ind. vech.  $v_1, v_2, \dots, v_k$  in  $\mathbb{R}^n$ , we can construct a set of mutually orthogonal vech.  $u_1, u_2, \dots, u_k$  with the same span. Formally:

$$\begin{cases} u_1 = v_1 \\ u_i = v_i - \text{proj}_{u_1}(v_i) - \text{proj}_{u_2}(v_i) - \dots - \text{proj}_{u_{i-1}}(v_i) \end{cases}$$

Ex: Apply GS-process to  $v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ ,  $v_3 = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}$ .

Sol:  $u_1 = v_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ .

$$\begin{aligned} u_2 &= v_2 - \text{proj}_{u_1}(v_2) = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{0}{3} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \end{aligned}$$

$$\begin{aligned} u_3 &= v_3 - \text{proj}_{u_2}(v_3) - \text{proj}_{u_1}(v_3) \\ &= \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{-1}{2} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} - \frac{4}{3} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 + \frac{1}{2} - \frac{4}{3} \\ 3 + 0 - \frac{4}{3} \\ 1 - \frac{1}{2} - \frac{4}{3} \end{pmatrix} = \begin{pmatrix} -\frac{5}{6} \\ \frac{5}{3} \\ -\frac{5}{6} \end{pmatrix} = \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \end{aligned}$$

Check:  $u_1 \cdot u_2 = 0$ ,  $u_1 \cdot u_3 = 0$ ,  $u_2 \cdot u_3 = 0$

$$u_1 \cdot u_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = 1 + 0 - 1 = 0$$

$$u_1 \cdot u_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} = \frac{5}{6} (-1 + 2 - 1) = \frac{5}{6} \cdot 0 = 0$$

$$u_2 \cdot u_3 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \cdot \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} = \frac{5}{6} (-1 + 0 + 1) = \frac{5}{6} \cdot 0 = 0 \quad \checkmark$$



Another check method: Note  $u \cdot v = u^T v$

$$\text{(i.e. } \begin{pmatrix} x \\ y \\ z \end{pmatrix} \cdot \begin{pmatrix} a \\ b \\ c \end{pmatrix} = (x \ y \ z) \begin{pmatrix} a \\ b \\ c \end{pmatrix} = [xa + yb + zc])$$

Take  $A = [u_1 | u_2 | u_3]$ , check

$$A^T A = \begin{bmatrix} u_1^T \\ u_2^T \\ u_3^T \end{bmatrix} [u_1 | u_2 | u_3] = \begin{bmatrix} u_1^T u_1 & u_1^T u_2 & u_1^T u_3 \\ u_2^T u_1 & u_2^T u_2 & u_2^T u_3 \\ u_3^T u_1 & u_3^T u_2 & u_3^T u_3 \end{bmatrix}$$

$$= \begin{bmatrix} u_1 \cdot u_1 & u_1 \cdot u_2 & u_1 \cdot u_3 \\ u_2 \cdot u_1 & u_2 \cdot u_2 & u_2 \cdot u_3 \\ u_3 \cdot u_1 & u_3 \cdot u_2 & u_3 \cdot u_3 \end{bmatrix} = \begin{bmatrix} |u_1|^2 & 0 & 0 \\ 0 & |u_2|^2 & 0 \\ 0 & 0 & |u_3|^2 \end{bmatrix}$$

if the  $u_i$ 's are mutually orthogonal...

Point:  $A^T A$  is a diagonal matrix if columns of  $A$  are mutually orthogonal... Should do: normalize the columns

of  $A$  (i.e. force  $|u_i| = 1$  for all  $i$  by taking suitable scalar multiples), then we obtain an "orthogonal matrix".

Defn: A matrix  $M$  is orthogonal when  $M^T = M^{-1}$  ( $M$  is  $n \times n$ ).

Prop:  $M$  is orthogonal if and only if the columns of  $M$  form an orthonormal basis for  $\mathbb{R}^n$ . pf: Easy exercise [2]

Defn: A basis of  $\mathbb{R}^n$  is orthonormal when the elements are mutually orthogonal and all have length 1.

Ex: Moment ago: we computed  $u_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $u_2 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}$ ,  $u_3 = \frac{5}{6} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix}$  form an orthogonal basis of  $\mathbb{R}^3$ . However,

$$|u_1| = \sqrt{1^2 + 1^2 + 1^2} = \sqrt{3}, \quad |u_2| = \sqrt{1+1} = \sqrt{2},$$

$$|u_3| = \frac{5}{6} \sqrt{1+4+1} = \frac{5}{6} \sqrt{6} = \frac{5}{\sqrt{6}}, \text{ so } B = \{u_1, u_2, u_3\} \text{ is}$$

not orthonormal. But

$$\begin{aligned} \hat{B} &= \left\{ \frac{1}{|u_1|} u_1, \frac{1}{|u_2|} u_2, \frac{1}{|u_3|} u_3 \right\} = \left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{5}{\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\} \\ &= \left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\} \stackrel{IS}{=} \text{orthonormal} \\ &= \left\{ \begin{pmatrix} \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \end{pmatrix}, \begin{pmatrix} \frac{1}{\sqrt{2}} \\ 0 \\ -\frac{1}{\sqrt{2}} \end{pmatrix}, \begin{pmatrix} -\frac{1}{\sqrt{6}} \\ \frac{2}{\sqrt{6}} \\ -\frac{1}{\sqrt{6}} \end{pmatrix} \right\} \end{aligned}$$

Check:  $\begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \end{bmatrix}^T \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \end{bmatrix} \stackrel{???}{=} I_3$

$$\begin{aligned} &\begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ -\frac{1}{\sqrt{6}} & \frac{2}{\sqrt{6}} & -\frac{1}{\sqrt{6}} \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & 0 & \frac{2}{\sqrt{6}} \\ \frac{1}{\sqrt{3}} & -\frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{6}} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{3} + \frac{1}{3} + \frac{1}{3} & \frac{1}{\sqrt{2}\sqrt{3}} + 0 - \frac{1}{\sqrt{2}\sqrt{3}} & \left[ -\frac{1}{\sqrt{3}\sqrt{6}} + \frac{2}{\sqrt{3}\sqrt{6}} - \frac{1}{\sqrt{3}\sqrt{6}} \right] \\ \frac{1}{\sqrt{2}\sqrt{3}} + 0 - \frac{1}{\sqrt{2}\sqrt{3}} & \frac{1}{2} + 0 + \frac{1}{2} & \frac{1}{\sqrt{2}\sqrt{6}} + 0 + \frac{1}{\sqrt{2}\sqrt{6}} \\ \left[ -\frac{1}{\sqrt{3}\sqrt{6}} + \frac{2}{\sqrt{3}\sqrt{6}} - \frac{1}{\sqrt{3}\sqrt{6}} \right] & \frac{1}{\sqrt{2}\sqrt{6}} + 0 + \frac{1}{\sqrt{2}\sqrt{6}} & \frac{1}{6} + \frac{4}{6} + \frac{1}{6} \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \checkmark \quad \text{☺} \end{aligned}$$

Remark: This will always couple an orthonormal basis from orthogonal one.

Algorithm (Extended Gram-Schmidt Process): Given  $v_1, v_2, \dots, v_k$  l.i. indep. in  $\mathbb{R}^n$ . To compute an orthonormal collection w/ same span:

- ① Apply the Gram-Schmidt Process to  $v_1, v_2, \dots, v_k$ .
- ② Normalize each output vector (i.e. scale each  $u_i$  by  $\frac{1}{|u_i|}$ ).

Ex: Apply extended GS process to  $V_1 = \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}$ ,  $V_2 = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}$ ,  $V_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ .  
 (NB: compare w/ previous example to note order matters for GS-process!)

Sol:  $u_1 = V_1 = \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}$

$$u_2 = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \text{proj}_{\begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix}}{\begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \\ 1 \end{pmatrix} - \frac{-1}{2} \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 \\ 3 \\ 1 \\ +0\frac{1}{2} \\ -\frac{1}{2} \end{pmatrix} = \begin{pmatrix} \frac{1}{2} \\ 3 \\ \frac{1}{2} \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}$$

$$u_3 = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \text{proj}_{\frac{1}{2} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \text{proj}_{\begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}}{\begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} - \frac{\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}}{\begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix} \cdot \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}} \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{8}{38} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} - \frac{0}{2} \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} - \frac{4}{19} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix} = \frac{1}{19} \begin{pmatrix} 19-4 \\ 19-24 \\ 19-4 \end{pmatrix} = \frac{1}{19} \begin{pmatrix} 15 \\ -5 \\ 15 \end{pmatrix} = \frac{5}{19} \begin{pmatrix} 3 \\ -1 \\ 3 \end{pmatrix}$$

$\text{proj}_{c\vec{u}}(\vec{v}) = \text{proj}_{\vec{u}}(\vec{v})$   
 Span is equal.

GS Process yields  $B = \left\{ \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{2} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}, \frac{5}{19} \begin{pmatrix} 3 \\ -1 \\ 3 \end{pmatrix} \right\}$ . Normalizing,

$$\hat{B} = \left\{ \frac{1}{\sqrt{2}} \begin{pmatrix} 0 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{38}} \begin{pmatrix} 1 \\ 6 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{19}} \begin{pmatrix} 3 \\ -1 \\ 3 \end{pmatrix} \right\}$$

$|\vec{c}\vec{u}| = |\vec{c}| |\vec{u}|$  so normalizing  $\vec{c}\vec{u} \mapsto \frac{1}{|\vec{c}|} \vec{c}\vec{u} = \frac{\vec{c}}{|\vec{c}| |\vec{u}|} \vec{u} = \frac{1}{|\vec{u}|} \vec{u}$   
 when  $\vec{c} \neq 0$

Q: Why are we computing orthonormal bases?

A: Orthonormal collections give vectors very nice representations...

In the GS process:  $u_i = v_i - \sum \text{proj}_{u_j}(v_i)$

$\leadsto v_i = \sum c_j u_j$

when  $u_j$ 's orthonormal

$\star$  So constant  $c_j u_j = \text{proj}_{u_j}(v_i) = \frac{u_j \cdot v_i}{u_j \cdot u_j} u_j = (u_j \cdot v_i) u_j$



Ex: The standard basis is an orthonormal basis.

$$v = \begin{pmatrix} a \\ b \\ c \end{pmatrix} = a e_1 + b e_2 + c e_3 = \underbrace{(v \cdot e_1)} e_1 + \underbrace{(v \cdot e_2)} e_2 + \underbrace{(v \cdot e_3)} e_3$$

Point: Orthonormal bases generalize the standard basis.

Ex: Compute  $\text{Rep}_{\hat{B}} \begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix}$  where  $\hat{B} = \left\{ \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \frac{1}{\sqrt{6}} \begin{pmatrix} -1 \\ 2 \\ -1 \end{pmatrix} \right\}$ .

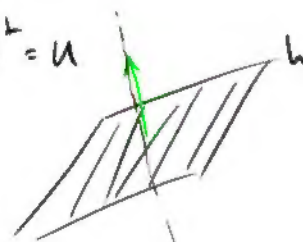
Sol:  $v = \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}$  has

$$\left. \begin{aligned} u_1 \cdot v &= \frac{1}{\sqrt{3}}(2+1+2) = \frac{5}{\sqrt{3}} \\ u_2 \cdot v &= \frac{1}{\sqrt{2}}(2+0-2) = 0 \\ u_3 \cdot v &= \frac{1}{\sqrt{6}}(-2+2-2) = -\frac{2}{\sqrt{6}} \end{aligned} \right\} \Rightarrow \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix} = \frac{5}{\sqrt{3}} u_1 + 0 u_2 - \frac{2}{\sqrt{6}} u_3$$



## ORTHOGONAL COMPLEMENTATION


Defn: A complement of subspace  $W \subseteq V$  is a subspace  $U$  such that every vector of  $V$  can be expressed uniquely as  $v = w + u$  where  $w \in W$  and  $u \in U$ .

Picture:  $W^\perp = U$    $W \subseteq \mathbb{R}^3$

Prop: If  $W \subseteq \mathbb{R}^n$ , then  $W^\perp = \{ u \in \mathbb{R}^n : u \cdot w = 0 \text{ for all } w \in W \}$  is the complement of  $W$ .

Proof: Every basis of  $W$  extends to a basis of  $\mathbb{R}^n$ .

Pick  $B$  a basis of  $W$ . Apply Extended GS to obtain  $\hat{B}$ .  $\hat{B}$  is still a basis of  $W$ . Extend to

$A = \hat{B} \cup \hat{D}$  a basis for  $\mathbb{R}^n$ .  $\hat{A} = \hat{B} \cup \hat{D}$ .  $W = \text{span}(\hat{B})$  and  $W^\perp = \text{span}(\hat{D})$ . 

Computationally: to compute  $W^\perp$ :

① express  $W = \text{Col}(A)$  for matrix  $A$ .

②  $W^\perp = \text{null}(A^T)$

Point  $\rightarrow$  Use  $A =$  matrix of any basis  $\smile$

